

Transforming Language Models into Smart Contract Audit Experts

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01

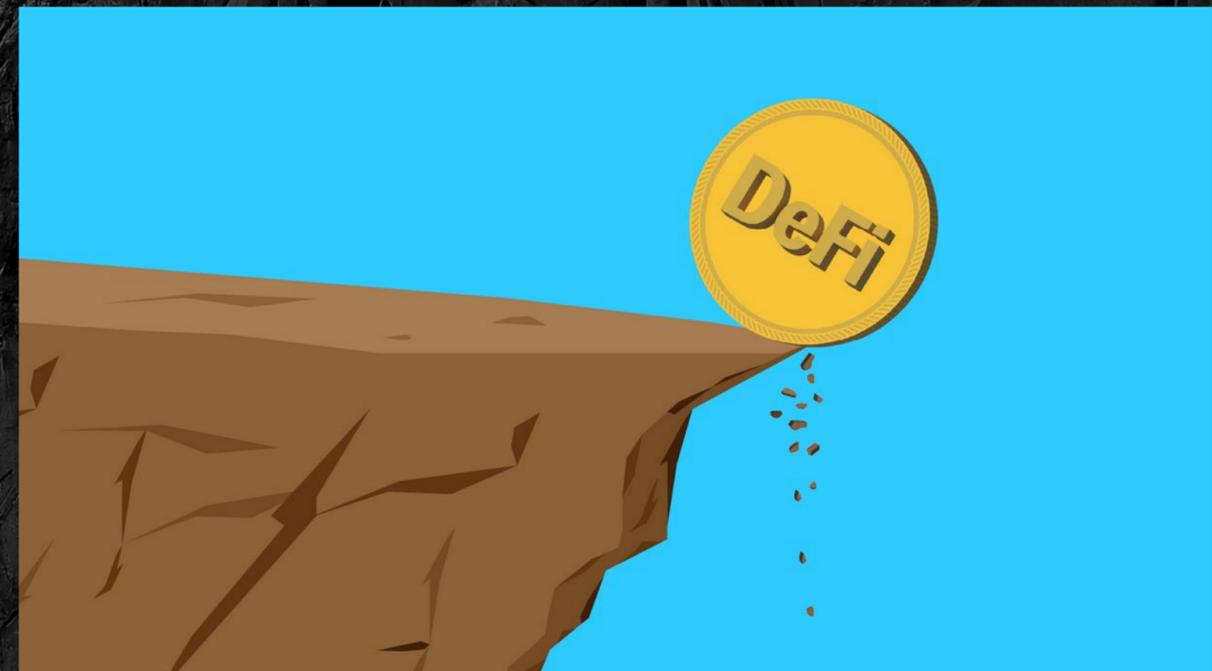
Introduction

Why traditional static analysis tools failed on smart contracts?

Prevalence of vulnerabilities on smart contracts

More than **80%** of the exploitable vulnerability for smart contracts are **machine undetectable**.

Most of them are related to the **business logic**.



Example for Logic Bugs

The first depositor could get all the shares and manipulate the price per share

To detect the vuln in the example:

1. Know it is a deposit
2. Find the share calculation statement
3. Check the if branch

```
1 function deposit(uint256 _amount) external returns (uint256) {
2     uint256 _pool = balance();
3     uint256 _before = token.balanceOf(address(this));
4     token.safeTransferFrom(msg.sender, address(this), _amount);
5     uint256 _after = token.balanceOf(address(this));
6     _amount = _after.sub(_before); // Additional check for deflationary
    tokens
7     uint256 _shares = 0;
8     if (totalSupply() == 0) {
9         _shares = _amount;
10    } else {
11        _shares = (_amount.mul(totalSupply())).div(_pool);
12    }
13    _mint(msg.sender, _shares);
14 }
```

Solution?



Challenges 1: Limited input length



GPT 3.5/4



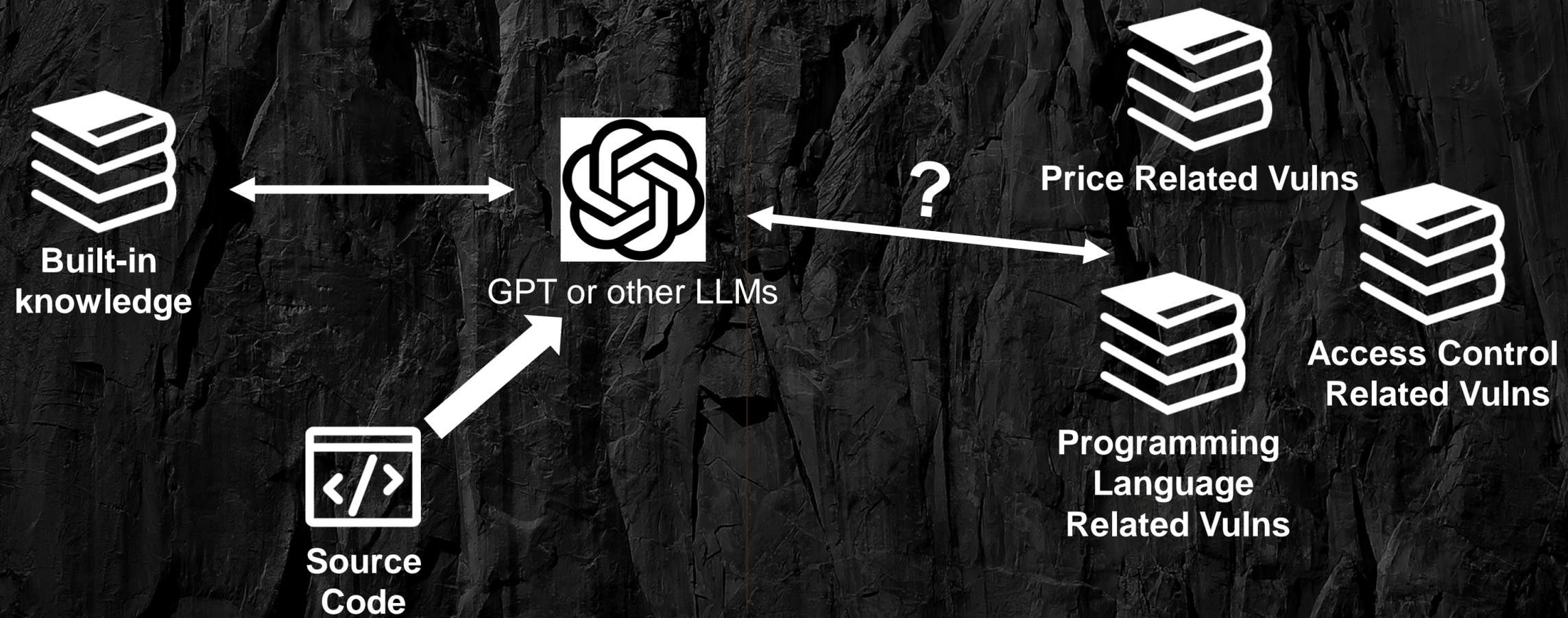
4K~32K token length



Real world projects are much longer



Challenges 2: Domain Specific Knowledge



Other Challenges

Too complex tasks



LLM may not be able to understand tasks that are too complex.

Hallucination



LLMs may have hallucinations and will not always give the correct answer.

Other problems



Lack of proper training data, and other challenges.



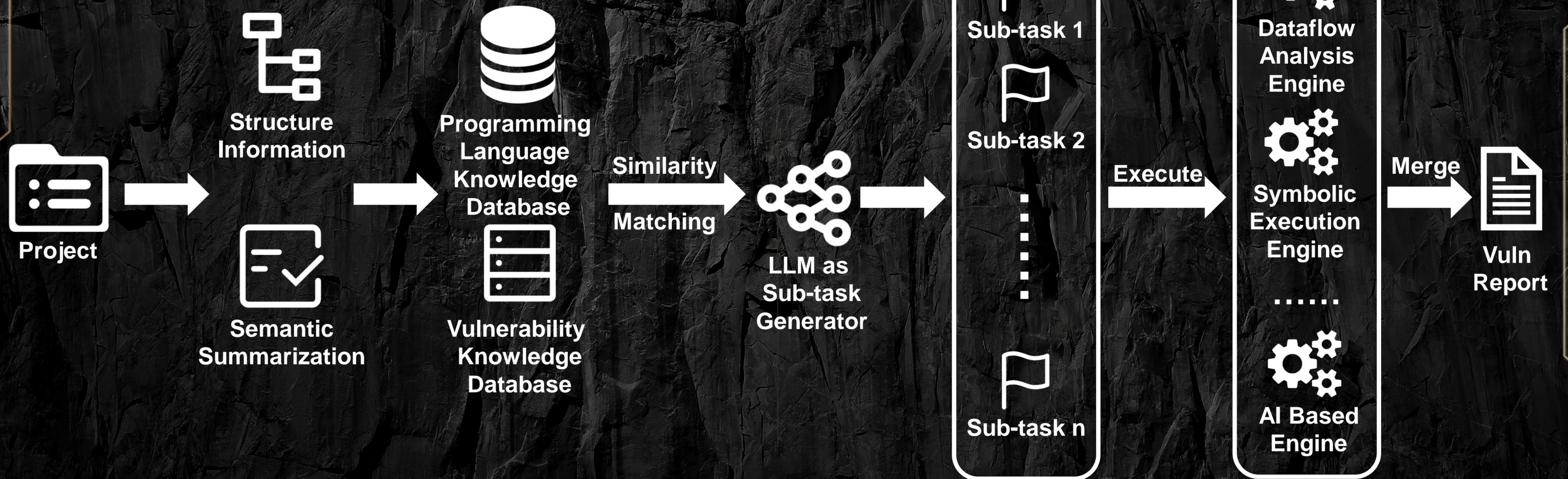
02

Method

How LLM could help detect such logic vulnerabilities?



AuditGPT



Step 1

Code Summarization & Knowledge Matching

Structural Information

- Call graph
- Usage of state variables
- Class inheritance
- Control flow
- ...

Semantic Information

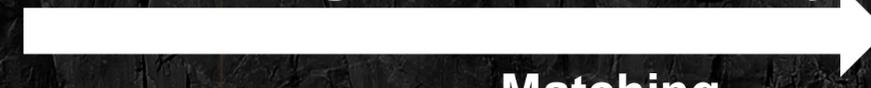
- Functionality of projects, contracts and functions
- Business model of the project

Find Related Code Segments



Code-Knowledge Pairs

1. Embedding 2. Vector Similarity



Matching



Vulnerability
Knowledge
Database



Step 2 Task Decomposing



Complex task

Neither handle by LLM or tools



Decompose



Simple Tasks

Can be handled by LLM or static analysis tools

[23:26:35] Running the analyzer

Retrieved Knowledge

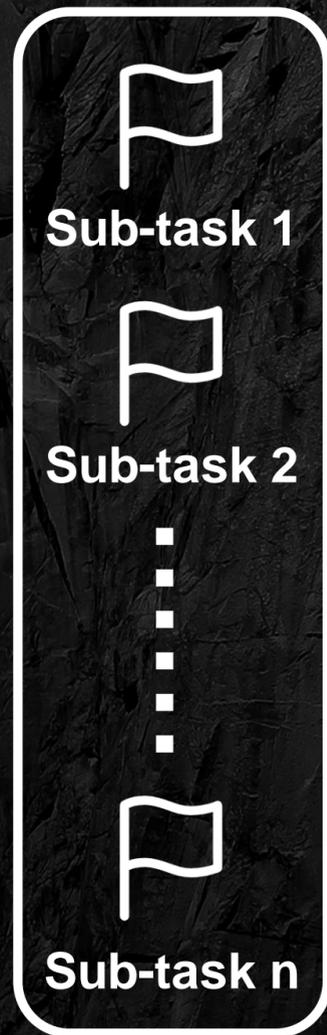
analyzer.py:17

When a pool allows the first depositor to lock an extremely small amount of liquidity, it can be manipulated to cause undesired outcomes for future liquidity providers, such as rounding down their minted shares to zero. By transferring a large amount of attributions to the pool after depositing a small amount, an attacker can obtain higher shares of the pool than they should. When the second provider deposits into the pool, the amount they get may round down to zero due to the small initial liquidity provided, effectively allowing the attacker to steal funds from other liquidity providers. The core vulnerability lies in the improper handling of the smallest possible amount deposited and attributions during the calculation of new liquidity provider shares. By exploiting this vulnerability, attackers can cause asset loss or compromise the pool's integrity.

Current Step

Does the deposit function calculate shares without checking for a minimum deposit amount?

Step 3 Sub-task Engine



Toolsets for
executing
the tasks

Current Step

Does the deposit function calculate shares without checking for a minimum deposit amount?

Dataflow Analysis

Source	Sink
<code>_amount</code>	<code>_shares</code>

[23:26:50] Step **Dataflow** finished with result *True* step.py:16

Dataflow Analysis Engine

Current Step

Does the function allow the deposit of extremely small amounts which can lead to extremely small or zero share allocation for future liquidity providers?

Symbolic Execution

Variable	Boundary
<code>_shares</code>	<code>1 ~ 100000000000000</code>

[23:26:55] Step **Symbolic Execution** finished with result *True* step.py:16

Symbolic Execution Engine

KnowLedge: have inside code statements that update/accrue interest/exchange rate, and have inside code statements that calculate/assign/distribute the balance/share/stake/fee/loan/reward

Code:

```
function pendingReward(address _user)
  external
  view
  returns (uint256)
{
  uint256 rewardShare = accRewardPerShare;
  uint256 staked = totalStaked;
  //if blockNumber is greater than the pool's 'lastRewardBlock' the pool's 'accRewardPerShare' is outdated,
  //we need to calculate the up to date amount to return an accurate reward value
  if (block.number > lastRewardBlock && staked > 0) {
    (rewardShare, ) = _computeUpdate();
  }
  return
  //rewards that the user had already accumulated but not claimed
  userPendingRewards[_user] +
  //subtracting the user's 'lastAccRewardPerShare' from the pool's 'accRewardPerShare' results in the amount of rewards per share
  //the pool has accumulated since the user's last claim, multiplying it by the user's shares results in the amount of new rewards claimable
  //by the user
  (balanceOf[_user] * (rewardShare - userLastAccRewardPerShare[_user])) /
  1e36;
}
```

Response

Yes.

AI Based Engine

03

Impact

How was the impact of AuditGPT in real world?



Vulnerability Bounties



32 new
vulnerabilities



9 new
vulnerabilities

TOTAL
30+
projects

TOTAL
10k
bounty per month



Demo

```
PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS 1 www.BANDICAM.COM  
→ src git:(main) X █  
  
█
```



Related Studies

AuditGPT is based on the following works:

1. GPTScan: Detecting Logic Vulnerabilities in Smart Contracts by Combining GPT with Program Analysis (ICSE 2024)
2. LLM4Vuln: A Unified Evaluation Framework for Decoupling and Enhancing LLMs' Vulnerability Reasoning (arXiv:2401.16185)
3. PropertyGPT: LLM-driven Formal Verification of Smart Contracts through Retrieval-Augmented Property Generation (arXiv:2405.02580)



GEEKCON

THANKS

